

Package: dynsbm (via r-universe)

May 21, 2026

Type Package

Title Dynamic Stochastic Block Models

Version 0.8

Date 2025-05-15

Maintainer Vincent Miele <vincent.miele@cnrs.fr>

Description Dynamic stochastic block model that combines a stochastic block model (SBM) for its static part with independent Markov chains for the evolution of the nodes groups through time, developed in Matias and Miele (2016) <doi:10.1111/rssb.12200>.

Imports Rcpp, parallel, RColorBrewer, grDevices, graphics, stats

LinkingTo Rcpp

License GPL-3

NeedsCompilation yes

Author Catherine Matias [aut], Vincent Miele [aut, cre]

Repository <https://vmielecnrs.r-universe.dev>

Date/Publication 2025-05-26 09:06:54 UTC

RemoteUrl <https://github.com/cran/dynsbm>

RemoteRef HEAD

RemoteSha 397e403d632dd0fcefeba79f3037864db5365370

Contents

dynsbm-package	2
adjacency.plot	2
antsMersch	4
connectivity.plot	5
foodwebWoodward	7
select.dynsbm	9
simdataT5Q4N40binary	12
simdataT5Q4N40continuous	12
simdataT5Q4N40discrete	13
sociopatternsPC2011	13

Index**15**

dynsbm-package	<i>Dynamic stochastic block model estimation</i>
----------------	--

Description

Estimation of a model that combines a stochastic block model (SBM) for its static part with independent Markov chains for the evolution of the nodes groups through time

Details

dynsbm is a R implementation of a model that combines a stochastic block model (SBM) for its static part with independent Markov chains for the evolution of the nodes groups through time. It deals with binary or weighted dynamic/temporal/evolving networks (with discrete or continuous edges).

Author(s)

Authors: Catherine Matias, Vincent Miele

Maintainer: Vincent Miele <vincent.miele@univ-lyon1.fr>

References

Catherine Matias and Vincent Miele, *Statistical clustering of temporal networks through a dynamic stochastic block model*, Journal of the Royal Statistical Society: Series B (2017) <http://dx.doi.org/10.1111/rssb.12200> <http://arxiv.org/abs/1506.07464>

Vincent Miele and Catherine Matias, *Revealing the hidden structure of dynamic ecological networks*, Royal Society Open Science (2017) <http://dx.doi.org/10.1098/rsos.170251> <https://arxiv.org/abs/1701.01355>

adjacency.plot	<i>Heatmap plot of the reorganized adjacency matrices associated to a dynamic stochastic block model.</i>
----------------	---

Description

Heatmap plot of the adjacency matrices with rows/columns reorganized according to the group membership associated to a dynamic stochastic block model.

Usage

```
adjacency.plot(dynsbm, Y, present=NULL, col=heat.colors(9))
```

Arguments

dynsbm	An object of class dynsbm retrieved with the function <code>select.dynsbm</code> .
Y	An object of class array of dimension (T x N x N) containing T adjacency matrices of size (N x N), where N is the number of nodes in the network and T is the number of time points.
present	NULL or an object of class matrix of size (N x T) containing the presence/absence (coded with 1/0 respectively) of each N nodes at each of the T time points. When set to NULL, this object is deduced from Y.
col	A list of colors such as that generated by <code>rainbow</code> , <code>heat.colors</code> , <code>topo.colors</code> , <code>terrain.colors</code> or similar functions.

Details

The T adjacency matrices are represented. The row/lines are reordered following the group membership (nodes of group 1 followed by nodes of group 2 and so on). Red lines correspond to group delineation.

The reordering is independent for each time step. The adjacency matrices do not contain the row/columns corresponding to absent nodes.

If `dynsbm` was estimated with `edge.type=="binary"`, the matrices cells are colored in white for value 0 or in the first color of the `col` argument vector for value 1. If `dynsbm` was estimated with `edge.type=="discrete"` or `edge.type=="continuous"`, the matrices cells are colored with a colored gradient for value >0.

Value

No return value, called for plotting.

Author(s)

Authors: Catherine Matias, Vincent Miele

Maintainer: Vincent Miele <vincent.miele@univ-lyon1.fr>

References

Catherine Matias and Vincent Miele, *Statistical clustering of temporal networks through a dynamic stochastic block model*, Journal of the Royal Statistical Society: Series B (2017) <http://dx.doi.org/10.1111/rssb.12200>
<http://arxiv.org/abs/1506.07464>

Vincent Miele and Catherine Matias, *Revealing the hidden structure of dynamic ecological networks*, Royal Society Open Science (2017) <http://dx.doi.org/10.1098/rsos.170251> <https://arxiv.org/abs/1701.01355>

Examples

```
#####
## 1 - binary case
data(simdataT5Q4N40binary)

## estimation for Q=1..5 groups
```

```

## better to use nstart>1 starting points
## but estimation can take 1-2 minutes
list.dynsbm <- select.dynsbm(simdataT5Q4N40binary,
  Qmin=1, Qmax=5, edge.type="binary", nstart=1)

## selection of Q=4
dynsbm <- list.dynsbm[[4]]

## plotting intra/inter connectivity patterns
adjacency.plot(dynsbm, simdataT5Q4N40binary)

#####
## 2 - continuous case
data(simdataT5Q4N40continuous)

## estimation for Q=1..5 groups
## better to use nstart>1 starting points
## but estimation can take 1-2 minutes
list.dynsbm <- select.dynsbm(simdataT5Q4N40continuous,
  Qmin=1, Qmax=5, edge.type="continuous", nstart=1)

## selection of Q=4
dynsbm <- list.dynsbm[[4]]

## plotting intra/inter connectivity patterns
adjacency.plot(dynsbm, simdataT5Q4N40continuous)

#####
## 3 - discrete case
data(simdataT5Q4N40discrete)

## estimation for Q=1..5 groups
## better to use nstart>1 starting points
## but estimation can take 1-2 minutes
list.dynsbm <- select.dynsbm(simdataT5Q4N40discrete,
  Qmin=1, Qmax=5, edge.type="discrete", K=4, nstart=1)

## selection of Q=4
dynsbm <- list.dynsbm[[4]]

## plotting intra/inter connectivity patterns
adjacency.plot(dynsbm, simdataT5Q4N40discrete)

```

Description

Colonies of the ant *Camponotus fellah* were followed with a tracking system that monitored the individual positions over days of observations and dynamic social interactions were deduced from

physical proximity. The data corresponds to a colony of $N=152$ ants observed during $T=10$ days. Edge weights were binned into $K=3$ categories corresponding to low, medium and high interaction intensity.

Usage

```
data(antsMersch)
```

Format

An array of size $10 \times 152 \times 152$.

Source

<http://datadryad.org/resource/doi:10.5061/dryad.8d8h7>

References

Mersch, D. P., Crespi, A., Keller, L., *Tracking individuals shows spatial fidelity is a key regulator of ant social organization*, *Science*, 340(6136), 1090-1093 (2013) <http://dx.doi.org/10.1126/science.1234316>

Miele, V and Matias, C, *Revealing the hidden structure of dynamic ecological networks*, Royal Society Open Science (2017)

Examples

```
data(antsMersch)

## better to use nstart>>1 starting points
## but estimation can be long;
## better to use nb.cores>1 cores
list.dynsbm <- select.dynsbm(antsMersch,
  Qmin=1, Qmax=6, edge.type="discrete", K=3,
  nstart=1, nb.cores=1)

## selection of Q=3 with the 'elbow' method
dynsbm <- list.dynsbm[[3]]

## plotting intra/inter connectivity patterns
connectivity.plot(dynsbm, antsMersch)
```

connectivity.plot	<i>Plot the connectivity characteristics between groups associated to a dynamic stochastic block model.</i>
-------------------	---

Description

Plot the connectivity characteristics between groups associated to a dynamic stochastic block model.

Usage

```
connectivity.plot(dynsbm, Y)
```

Arguments

dynsbm An object of class `dynsbm` retrieved with the function `select.dynsbm`.

Y An object of class `array` of dimension $(T \times N \times N)$ containing T adjacency matrices of size $(N \times N)$, where N is the number of nodes in the network and T is the number of time points.

Details

Interaction presence and intensity between nodes in any of the groups to the others are represented in a $Q \times Q$ matrix. The cell in line q /column l deals with the connectivity between groups q/l . Each cell contains a curve with T time points on the x -axis corresponding to the T proportions of present edges over all the possible edges, where Q is the number of groups and T is the number of time points, and If `dynsbm` was estimated with `edge.type=="binary"`, the area below the curve is filled in light blue. If `dynsbm` was estimated with `edge.type=="discrete"`, the area below the curve is divided into K areas corresponding to the proportion of edges with value 1 to K (the darker blue, the greater edge intensity). If `dynsbm` was estimated with `edge.type=="continuous"`, the area below the curve is filled with a colored gradient representing the mean edge intensity (the darker blue, the greater).

Value

No return value, called for plotting.

Author(s)

Authors: Catherine Matias, Vincent Miele
 Maintainer: Vincent Miele <vincent.miele@univ-lyon1.fr>

References

Catherine Matias and Vincent Miele, *Statistical clustering of temporal networks through a dynamic stochastic block model*, Journal of the Royal Statistical Society: Series B (2017) <http://dx.doi.org/10.1111/rssb.12200>
<http://arxiv.org/abs/1506.07464>

Vincent Miele and Catherine Matias, *Revealing the hidden structure of dynamic ecological networks*, Royal Society Open Science (2017) <http://dx.doi.org/10.1098/rsos.170251> <https://arxiv.org/abs/1701.01355>

Examples

```
#####
## 1 - binary case
data(simdataT5Q4N40binary)

## estimation for Q=1.5 groups
## better to use nstart>1 starting points
## but estimation can take 1-2 minutes
```

```

list.dynsbm <- select.dynsbm(simdataT5Q4N40binary,
  Qmin=1, Qmax=5, edge.type="binary", nstart=1)

## selection of Q=4
dynsbm <- list.dynsbm[[4]]

## plotting intra/inter connectivity patterns
connectivity.plot(dynsbm, simdataT5Q4N40binary)

#####
## 2 - continuous case
data(simdataT5Q4N40continuous)

## estimation for Q=1..5 groups
## better to use nstart>1 starting points
## but estimation can take 1-2 minutes
list.dynsbm <- select.dynsbm(simdataT5Q4N40continuous,
  Qmin=1, Qmax=5, edge.type="continuous", nstart=1)

## selection of Q=4
dynsbm <- list.dynsbm[[4]]

## plotting intra/inter connectivity patterns
connectivity.plot(dynsbm, simdataT5Q4N40continuous)

#####
## 3 - discrete case
data(simdataT5Q4N40discrete)

## estimation for Q=1..5 groups
## better to use nstart>1 starting points
## but estimation can take 1-2 minutes
list.dynsbm <- select.dynsbm(simdataT5Q4N40discrete,
  Qmin=1, Qmax=5, edge.type="discrete", K=4, nstart=1)

## selection of Q=4
dynsbm <- list.dynsbm[[4]]

## plotting intra/inter connectivity patterns
connectivity.plot(dynsbm, simdataT5Q4N40discrete)

```

Description

This dataset concerns the aquatic macro-invertebrate community of Broadstone Stream in south-east England. Six seasonal connectance food webs were recorded, one every two months from May 1996 to April 1997. We restricted here to simple presence/absence information on species (nodes) and binary feeding links (edges). This dataset forms a dynamic trophic network with T=6 snapshots

(May, August, October, December 1996, February, April 1997). Five species were not sampled each month. Each network is directed.

Usage

```
data(foodwebWoodward)
```

Format

An array of size 6x26x26.

Source

Table 2 of Woodward et al

References

Woodward, G., Speirs, D. C., Hildrew, A. G., *Quantification and Resolution of a Complex, Size-Structured Food Web. In Food Webs: From Connectivity to Energetics*, Vol. 36 of Advances in Ecological Research, pp. 85-135 Academic Press (2005) [http://dx.doi.org/10.1016/S0065-2504\(05\)36002-8](http://dx.doi.org/10.1016/S0065-2504(05)36002-8)

Miele, V and Matias, C, *Revealing the hidden structure of dynamic ecological networks*, Royal Society Open Science (2017)

Examples

```
data(foodwebWoodward)

## mandatory to use many nstart>>1 starting points
## but estimation can be long;
## better to use nb.cores>1 cores
list.dynsbm <- select.dynsbm(Y=foodwebWoodward$Y,
  present=foodwebWoodward$present,
  Qmin=1, Qmax=6, edge.type="binary",
  directed=TRUE, self.loop=TRUE,
  nstart=200, nb.cores=1)

## selection of Q=4 with the ICL method
dynsbm <- list.dynsbm[[4]]

## plotting intra/inter connectivity patterns
connectivity.plot(dynsbm, foodwebWoodward$Y)
```

select.dynsbm	<i>Dynamic stochastic block model estimation for different number of groups.</i>
---------------	--

Description

Estimation of dynamic stochastic block models for different number of groups. Each model combines a stochastic block model (SBM) for its static part with independent Markov chains for the evolution of the nodes groups through time.

Usage

```
select.dynsbm(Y, present=NULL, Qmin, Qmax,
  edge.type=c("binary","discrete","continuous"), K=-1,
  directed=FALSE, self.loop=FALSE,
  nb.cores=1,
  iter.max=20, nstart=25, perturbation.rate=0.2,
  fixed.param=FALSE, bipartition=NULL,
  plot=TRUE)
```

Arguments

Y	An object of class <code>array</code> of dimension (T x N x N) containing T adjacency matrices of size (N x N), where N is the number of nodes in the network and T is the number of time points.
present	NULL or an object of class <code>matrix</code> of size (N x T) containing the presence/absence (coded with 1/0 respectively) of each N nodes at each of the T time points. When set to NULL, this object is deduced from Y (see the "Details" section). Any node must be present at least once among the time points.
Qmin	Minimum number of groups >1.
Qmax	Maximum number of groups.
edge.type	Type of adjacency matrices. This should be (an unambiguous abbreviation of) one of <code>binary</code> , <code>discrete</code> or <code>continuous</code> . See the "Details" section.
K	Only if <code>edge.type=="discrete"</code> . Number of non-zero discrete values (i.e. in {1,...,K}).
directed	If TRUE, the network is supposed to be directed (and therefore Y is supposed to be asymmetric).
self.loop	If TRUE, self-loops (edges from one node to the same node) are allowed and taken into account in the estimation procedure.
nb.cores	Number of cores to use, i.e. how many child processes and how many threads will be run simultaneously during the initialization and the estimation steps respectively.
iter.max	Maximal number of algorithm iterations.

nstart	Number of starting points for the iterative estimation algorithm. See the "Details" section.
perturbation.rate	Rate of perturbation (in [0,1], see nstart) for the iterative estimation algorithm. This rate is the fraction of nodes for which its group is randomly shuffled.
fixed.param	If TRUE, the model parameters remain fixed and constant in time. By default, fixed.param is automatically set to TRUE in the bipartite case (i.e. bipartition is not NULL; see the "Details" section).
bipartition	NULL or a vector of size N specifying a node bipartition in the case of bipartite networks (see the "Details" section). Each element of this vector is set to 1 or 2 to specify the node belongs to the first or second set of nodes.
plot	Display a plot with the loglikelihood and the ICL criteria if edge.type=="binary"/"continuous". See the "Details" section.

Details

This function deals with binary or weighted dynamic/temporal/evolving networks (with discrete or continuous edges). The adjacency matrices must be coded with 0/1 in the binary case, with 0/y where y belongs to the set {1,...,K} in the discrete case or with 0/y where y is numeric, must be positive and is supposed to fit a gaussian mixture in the continuous case.

Presence/absence information allows to model node's arrival or departure, birth or death, or simply enables to specify missing data (as absent nodes). If this information is missing (NULL), the presence/absence is deduced automatically from Y by searching for nodes that do not participate in any edges (lines/columns of O in Y) and declaring them as absent. This function does not support the existence of nodes that are never present (error message in this case).

The estimation algorithm is iterative and rely on a starting point. Therefore, it is possible to start the algorithm many times with 'nstart' starting points. The first starting point is obtained with an ad-hoc use of the kmeans function. The following starting point are obtained by perturbing the first one (see perturbation.rate). The greater nstart, the more accurate the results.

To select the best number of groups, the "elbow" method consists in finding the point where the slope of the loglikelihood is decreasing (i.e. the loglikelihood is reaching a plateau). If edge.type=="binary", the ICL criteria (plotted in red) has to be used : the best number of nodes is supposed to maximize the ICL criteria.

This function has been extended to the case of bipartite networks. In this case, despite Y has to be of dimension (T x N x N), it is possible to give a bipartition of the nodes into two disjoint sets. For statistical reasons, fixed.param is automatically set to TRUE. Given the total number of groups Q between Qmin and Qmax, there is Q/2 groups for each set of nodes (when Q is odd, there is floor(Q/2)+1 groups for the largest set of nodes); however, there is no guaranty that the final groups are coherent with the bipartition, i.e. that any group is composed by nodes of one of the two sets (if not, a warning message is generated).

Value

Returns a list of dynsbm objects. Each object of class dynsbm is a list with the following components:

transition The Markov chain transition matrix of size (Q x Q).

membership	An object of class <code>matrix</code> of size (N x T) containing the group membership estimated by MAP (>0, =0 for absent nodes).
beta	An object of class <code>matrix</code> of size (T x Q x Q) containing the sparsity parameters of the model.
gamma	Only if <code>edge.type=="discrete"</code> . An object of class <code>matrix</code> of size (T x Q x Q x K) containing the model parameters.
mu/sigma	Only if <code>edge.type=="continuous"</code> . An object of class <code>matrix</code> of size (T x Q x Q) and a vector containing the model parameters.
loglikelihood	Completed data log-likelihood.
iter	Number of used algorithm iterations.
directed	Specifies whether the model is build for directed networks.
self.loop	Specifies whether the model allows self-loops.

Author(s)

Authors: Catherine Matias, Vincent Miele

Maintainer: Vincent Miele <vincent.miele@univ-lyon1.fr>

References

Catherine Matias and Vincent Miele, *Statistical clustering of temporal networks through a dynamic stochastic block model*, Journal of the Royal Statistical Society: Series B (2017) <http://dx.doi.org/10.1111/rssb.12200> <http://arxiv.org/abs/1506.07464>

Vincent Miele and Catherine Matias, *Revealing the hidden structure of dynamic ecological networks*, Royal Society Open Science (2017) <http://dx.doi.org/10.1098/rsos.170251> <https://arxiv.org/abs/1701.01355>

Examples

```
data(simdataT5Q4N40binary)

## estimation for Q=1..6 groups
## better to use nstart>1 starting points
## but estimation can take 1-2 minutes
list.dynsbm <- select.dynsbm(simdataT5Q4N40binary,
  Qmin=1, Qmax=6, edge.type="binary", nstart=1)

## selection of Q=4
dynsbm <- list.dynsbm[[4]]

## plotting intra/inter connectivity patterns
connectivity.plot(dynsbm, simdataT5Q4N40binary)
```

simdataT5Q4N40binary *Dataset generated by a dynamic stochastic block model - binary case*

Description

5 snapshots of a network with 40 nodes with binary weights, generated by a dynamic stochastic block model with 4 groups.

Usage

```
data(simdataT5Q4N40binary)
```

Format

An array of size 5x40x40.

simdataT5Q4N40continuous
Dataset generated by a dynamic stochastic block model - continuous case

Description

5 snapshots of a network with 40 nodes with continuous weights, generated by a dynamic stochastic block model with 4 groups.

Usage

```
data(simdataT5Q4N40continuous)
```

Format

An array of size 5x40x40.

`simdataT5Q4N40discrete`*Dataset generated by a dynamic stochastic block model - discrete case*

Description

5 snapshots of a network with 40 nodes with discrete weights ($K=3$ categories), generated by a dynamic stochastic block model with 4 groups.

Usage

```
data(simdataT5Q4N40discrete)
```

Format

An array of size 5x40x40.

`sociopatternsPC2011`*High school dynamic contact networks in the PC class in 2011 - Sociopatterns collaboration*

Description

The dataset consists in face-to-face encounters of high school students of a class from a French high school. In this class called 'PC', interactions were recorded during 4 days (Tuesday to Friday) in Dec. 2011. We kept only the 27 (out of 31) students that appear every day. Interaction times were aggregated by days to form a sequence of 4 different networks. These are undirected and weighted networks, the weight of an interaction between two individuals being the number of interactions between these 2 individuals divided by the number of time points for which at least two individuals interacted; we discretize these data into 3 bins.

We sincerely acknowledge the SocioPatterns collaboration for providing the dataset (<http://www.sociopatterns.org>)

Usage

```
data(sociopatternsPC2011)
```

Format

An array of size 4x27x27.

Source

<http://www.sociopatterns.org/datasets/high-school-dynamic-contact-networks/>

References

Julie Fournet and Alain Barrat, *Contact patterns among high school students*, PLoS ONE 9(9):e107878 (2014).

Catherine Matias and Vincent Miele, *Statistical clustering of temporal networks through a dynamic stochastic block model*, Journal of the Royal Statistical Society: Series B (2016) <http://dx.doi.org/10.1111/rssb.12200>
<http://arxiv.org/abs/1506.07464>
<http://www.sociopatterns.org/>

Index

* **cluster**

dynsbm-package, [2](#)

* **package**

dynsbm-package, [2](#)

adjacency.plot, [2](#)

antsMersch, [4](#)

connectivity.plot, [5](#)

dynsbm (dynsbm-package), [2](#)

dynsbm-package, [2](#)

foodwebWoodward, [7](#)

select.dynsbm, [9](#)

simdataT5Q4N40binary, [12](#)

simdataT5Q4N40continuous, [12](#)

simdataT5Q4N40discrete, [13](#)

sociopatternsPC2011, [13](#)